A global analysis of the relationship between violence and life expectancy Big Data in Social Sciences

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Introduction

Measuring violence in a country is a challenging task.

- ▶ Violence is a multifaceted concept
- lt depends on many different factors
- ▶ Multiple sources are needed to capture its complexities

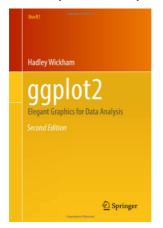
Research Question

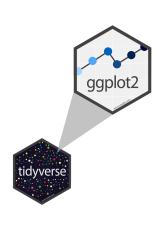
- ▶ The level of violence may affect the mortality of a country
- ▶ Today, we will rely on two major data sources to explore the relationship between violence and mortality:
 - Data on Violece from the Institute for Economics & Peace
 - Data on Mortality from the United Nations
- ▶ We will produce multiple descriptive plots with functions from ggplot

What about ggplot?

ggplot2 is an R package for producing statistical, or data, graphics

It is already available in tidyverse!





ggplot

Major advantages:

- creating graphs by combining independent components
- detailed theming system to generate nice-looking graphs
- intuitive grammar
- praphs are R objects

Indicators for violence and mortality

- Violence → Global Peaceful Index (GPI) that measures the violence of a country across three dimensions:
 - ongoing domestic and international conflict
 - societal safety and security
 - militarization
- ▶ Mortality \rightarrow life expectancy at birth (e_0)
 - common indicator for capturing the mortality of a population

Data files

Two .txt data files on Virtuale

- $ightharpoonup gpi_data.txt
 ightarrow GPI$ for multiple countries over 2008-2022
- Iife_exp_data.txt → life expectancy estimates for multiple countries over 2008-2022 couple
 - Regional classification included
 - Total population size

Upload data in Rstudio

Upload tidyverse and the data sets

```
#install.packages('ggthemes')
#install.packages('RColorBrewer')
library('RColorBrewer') # various qualitative color palettes
library('ggthemes') # various themes in ggplot
library("tidyverse") # ggplot2

gpi_data <- read.table('Data/gpi_data.txt',header=T)
life_exp_data <- read.table('Data/life_exp_data.txt',header=T)</pre>
```

Combine the two data sets by iso3 (unique to each country) and Year

```
data <- inner_join(gpi_data,life_exp_data,by=c('iso3','Year'))</pre>
```

We match all records in *gpi_data*, whose *iso3* and *Year* values have a correspondence in *life_exp_data*

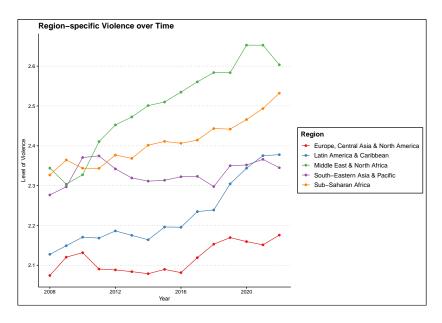
The non-matching records, either in <code>gpi_data</code> or in <code>life_exp_data</code>, are dropped

Create a plot displaying the evolution of violence over the time period 2008-2022 by world region

Calculate region- and year-specific levels of violence

```
data_violence_region = data |>
    group_by(Year,area) |>
    summarize(GPI=weighted.mean(x=GPI,w=pop,na.rm = FALSE))
```

Generate the plot



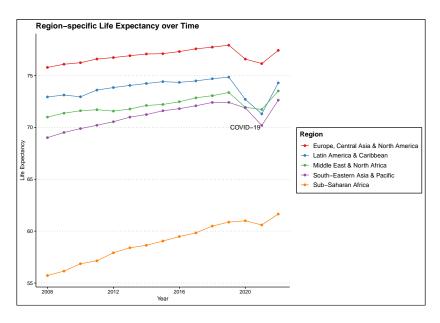
Create a similar visual aid for life expectancy

Calculate region-specific life expectancy estimates

```
data_life_exp_region = data |>
    group_by(Year,area) |>
    summarize(e0=weighted.mean(x=e0,w=pop,na.rm = FALSE))
```

Generate the plot

```
plot2 = ggplot(data=data_life_exp_region, # data input
mapping=aes(x=Year,y=e0,color=area))+ # relationships
geom_line()+ # add a line for each country
geom_point()+ # add points
annotate("text",x=2020,y=70,label="COVID-19")+ # add text
theme_clean()+ # specify a theme
xlab('Year')+ #label for title of x-axis
ylab('Life Expectancy') + # label for title of y-axis
scale_color_brewer(name = "Region", palette = "Set1")+ # change color palette
ggtitle('Region-specific Life Expectancy over Time') # title
```



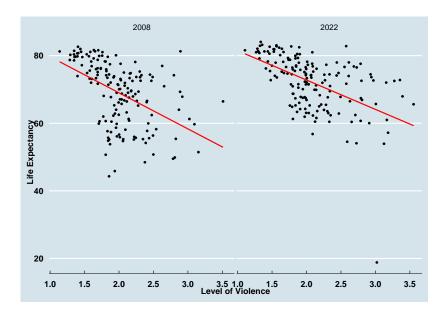
Produce two scatter plots to display the relationship between violence and life expectancy in 2008 and 2022. Add also a regression line.

Select records for years 2008 and 2022

```
data_scatter = data |>
filter(Year %in% c(2008,2022))
```

Generate the plot

```
plot3 = ggplot(data=data_scatter, # data input
mapping=aes(x=GPI,y=e0)) + # relationships
geom_point() + # scatter plot
xlab('Level of Violence') + # x-axis label
ylab('Life Expectancy') + # y-axis label
theme_economist() + # pick a background
geom_smooth(method = "lm", se = FALSE,color='red') + # regression line
facet_wrap(~Year) + # seperate (sub)plot by year
theme(axis.text.y = element_text(size=15,face="bold"), # font of y-axis text
axis.title.y = element_text(size=15,face="bold"), # font of y-axis title
axis.text.x = element_text(size=15,face="bold"),
axis.title.x = element_text(size=15,face="bold")) # font of x-axis title
```



Display the distribution of life expectancy in the 20 most violent countries and in the 20 most peaceful countries in 2022

Let's create the data sets Most violent countries

```
data_most_violence = data |>
    filter(Year==2022) |>
    slice_max(GPI,n=20) |>
    mutate(label='Most Violent')
```

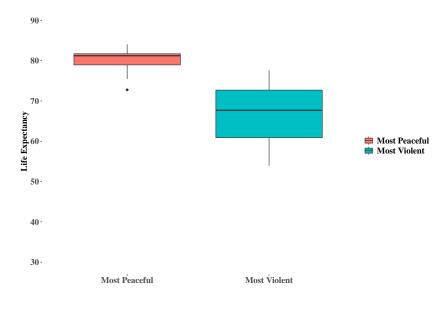
Most peaceful countries

```
data_most_peaceful = data |>
filter(Year==2022) |>
slice_min(GPI,n=20) |>
mutate(label='Most Peaceful')
```

combine the two data sets by row

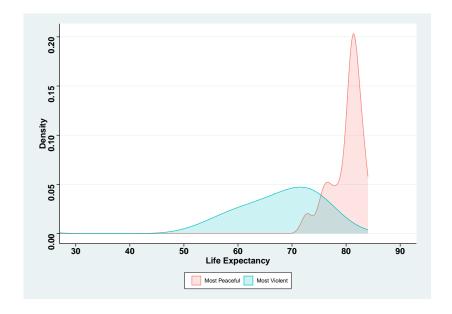
```
data_plot=rbind(data_most_violence,data_most_peaceful)
```

```
plot4=ggplot(data=data_plot, # data
mapping=aes(x=label,y=e0,fill=label))+ # relationships
geom_boxplot()+ # boxplot
coord cartesian(ylim=c(30,90))+ # fix y-axis limits
scale_y_continuous(breaks=seq(30,90,10),
                   labels=seq(30,90,10))+ # fix y-axis labels
theme_tufte()+ # pick background
scale fill discrete(name='')+ # drop name from legend
xlab('')+ # no name for x-axis title
ylab('Life Expectancy')+ # y-axis title
theme(axis.text.y = element_text(size=15, face="bold"), # font of y-axis text
legend.text = element_text(size=15,face="bold"), # legend text details
axis.title.y = element text(size=15, face="bold"), # font of y-axis title
axis.text.x = element_text(size=15,face="bold"),  # text of x-axis title
axis.title.x = element text(size=15, face="bold")) # font of x-axis title
```



Perform the same task using a different visual aid.

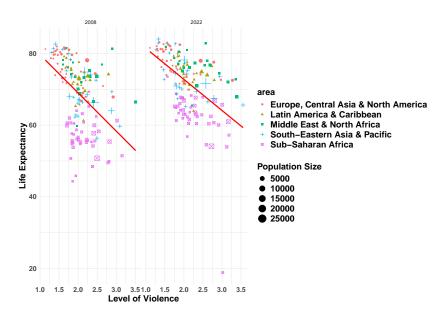
```
plot5 = ggplot(data=data_plot, # data input
aes(x=e0,color=label,fill=label))+ # relationships
geom density(alpha = 0.2, na.rm = TRUE) + # density
theme_stata()+ # background
scale_fill_discrete(name='')+ # no title to legend
scale_color_discrete(name='')+ # no title to legend
coord_cartesian(xlim=c(30,90))+ # fix x-axis limits
scale_x_continuous(breaks=seq(30,90,10),
                  labels=seq(30,90,10))+ # control x-axis labels
ylab('Density')+ # y-axis title
xlab('Life Expectancy')+ # x-axis title
theme(axis.text.y = element_text(size=15, face="bold"), # font of y-axis text
legend.position = "bottom", # legend position
axis.title.y = element text(size=15, face="bold"), # font of y-axis title
axis.text.x = element_text(size=15,face="bold"), # font of x-axis text
axis.title.x = element text(size=15, face="bold")) # font of x-axis title
```



Produce a scatter plot to display the relationship between violence and life expectancy in 2008 and 2022. Make sure to set different shapes and colors for the points

according to the region where they are located. Fix the size of the points according to the population size of the country

```
plot6 = ggplot(data=data_scatter, # data input
mapping=aes(x=GPI,y=e0))+ # relationships
geom_point(aes(shape=area,color=area,size=pop))+ # scatter plot
xlab('Level of Violence')+ # label x-axis
vlab('Life Expectancy')+ # label y-axis
theme_minimal()+ # plot them
geom_smooth(method = "lm", se = FALSE,color='red')+ # regression line
labs(size = 'Population Size')+ # legend label
facet_wrap(~Year) + # seperate by year
theme(axis.text.y = element_text(size=12, face="bold"), # font of y-axis text
axis.title.y = element_text(size=15, face="bold"), # font of y-axis title
legend.title = element_text(size=15, face="bold"), # font of legend title
legend.text = element text(size=15,face="bold"), # font of legend text
axis.text.x = element_text(size=12,face="bold"),
axis.title.x = element text(size=15, face="bold")) # font of x-axis title
```



Display the top 5 countries with the largest increase and the top 5 with largest decrease in violence over the period 2008-2022 in Europe, Central Asia & North America $\,$

```
data_high = data |>
filter(Year %in% c(2008,2022)) |>
select(Year,Country,area,GPI) |>
pivot_wider(names_from='Year',values_from='GPI') |>
mutate(var=(`2022`-`2008`)/`2008`) |>
group_by(area) |>
slice max(var,n=5) |>
mutate(label='Highest Violence Increase')
data low = data |>
filter(Year %in% c(2008,2022)) |>
select(Year,Country,area,GPI) |>
pivot_wider(names_from='Year',values_from='GPI') |>
mutate(var=(^2022`-^2008`)/^2008`) |>
group_by(area) |>
slice min(var,n=5) |>
mutate(label='Highest Violence Decrease')
```

```
plot7 = rbind(data_high,data_low) |>
filter(area == 'Europe, Central Asia & North America') |>
ggplot(aes(x=Country,y=var,fill=label))+ # data
geom bar(stat='identity') + # bar plot
theme_clean()+ # background
xlab('Country')+ # x-axis title
ylab('Variation in Violence')+ # y-axis title
scale_fill_brewer(name='',palette='Set2')+ # color palette
scale y continuous(labels = scales::percent)+ # y-axis labels
theme(axis.text.x = element_text(angle = 45,
vjust = 1, hjust = 1, size = 12, face = "bold"), # x-axis text
legend.position = 'bottom', # legend position
axis.text.y = element_text(size = 12, face = "bold"), # y-axis text
axis.title.x = element text(size = 15, face = "bold"), # x-axis title
axis.title.y = element_text(size = 15, face = "bold"), # y-axis title
legend.text = element_text(size = 15, face = "bold")) # legend text
```

